1 Evidence of Convective Redistribution of Carbon Monoxide in Aura Tropospheric 2 Emission Sounder (TES) and Microwave Limb Sounder (MLS) Observations 3 4 M. Manyin, A. R. Douglass, M. R. Schoeberl 5 6 Michael Manyin 7 Science Systems and Applications, Inc., Lanham, Maryland, USA 8 Email: Michael.Manyin@nasa.gov 9 Address: NASA/GSFC 10 Laboratory for Atmospheres 11 12 Code 613.3 Greenbelt, Maryland 20771 13 14 15 Anne Douglass 16 NASA Goddard Space Flight Center, Greenbelt, Maryland, USA 17 Email: Anne.R.Douglass@nasa.gov 18 Address: 19 NASA/GSFC 20 Laboratory for Atmospheres 21 Code 613.3 22 Greenbelt, Maryland 20771 23 24 Mark Schoeberl 25 Science and Technology Corporation, Hampton, Virginia, USA 26 Email: Mark.Schoeberl@mac.com 27 Address: 28 10005 Old Columbia Rd. 29 Suite M-150 Columbia, MD 21046 30 31

31 Abstract

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Vertical convective transport is a key element of the tropospheric circulation. Convection lofts air from the boundary layer into the free troposphere, allowing surface emissions to travel much further, and altering the rate of chemical processes such as ozone production. This study uses satellite observations to focus on the convective transport of CO from the boundary layer to the mid and upper troposphere. Our hypothesis is that strong convection associated with high rain rate regions leads to a correlation between mid level and upper level CO amounts. We first test this hypothesis using the Global Modeling Initiative (GMI) chemistry and transport model. We find the correlation is robust and increases as the precipitation rate (the strength of convection) increases. We next examine three years of CO profiles from the Tropospheric Emission Sounder (TES) and Microwave Limb Sounder (MLS) instruments aboard EOS Aura. Rain rates are taken from the Tropical Rainfall Measuring Mission (TRMM) 3B-42 multi-satellite product. Again we find a correlation between mid-level and upper tropospheric CO, which increases with rain rate. Our result shows the critical importance of tropical convection in coupling vertical levels of the troposphere in the transport of trace gases. The effect is seen most clearly in strong convective regions such as the Intertropical Convergence Zone.

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51 1. Introduction/Background

Convection plays a vital role in tropospheric circulation as well as the distribution of trace gases. Surface level trace gas emissions are usually mixed within the atmospheric boundary layer (BL) via turbulent diffusion but are frequently blocked from entering the free troposphere (FT) by the inversion layer at the top of the BL. Convection, along with weather fronts and orographic processes, overcome the BL inversion trapping, moving air into the FT [*Liu et al.*, 2003]. Deep convection reaches the tropopause and sometimes even overshoots into the lower stratosphere. *Cotton et al.* [1995] estimated that convective (cloud related) processes vent the mass of the boundary layer tens of times over the course of a year.

Once convection lofts constituents into the FT, they can be transported intercontinental distances because of the consistent large scale flow in that region. Convection also contributes to inter-hemispheric transport: strong convergence from the Intertropical Convergence Zone (ITCZ) convective systems can draw air from both hemispheres and loft it high enough to be pulled into the Hadley circulation. In addition to these effects on transport, convection can also dramatically alter the rate of chemical processes because of its speed. Convective time scales are very short, on the order of hours for surface to UT, as opposed to days for synoptic scale processes. The rapid transport of chemical species means that short-lived species can still reach the upper troposphere. For example, when urban air with NO_x and other ozone precursors is quickly lofted to the FT, ozone production is greatly enhanced [*Pickering*, 1992] because the transport is fast compared to the NO_x lifetime. Convective transport also extends the lifetime of some species by

reducing their exposure to loss processes in the lower troposphere; this expands their

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76 range of chemical influence [Dickerson et al., 1987]. 77 78 Convective processes are complex, involving surface radiative heating, updrafts from 79 lower tropospheric levels, detrainment at heights up to and above the tropopause, 80 condensation, latent heat release, precipitation, downdrafts and evaporative cooling (e.g. 81 Gamache [1982]). Other factors include melting ice particles, and turbulent mixing due to 82 vertical shear and gravity wave instabilities. The scale of convective systems can range 83 from isolated thunderstorms up to mesoscale complexes measuring 1000+ km. The 84 wide-ranging impact of convection on chemical transport makes it crucial for us to 85 quantify convective transport and evaluate its representation in models. 86 87 Previous studies of convection have focused on various features, including (i) showing 88 evidence of convective transport from BL to upper troposphere/lower stratosphere region 89 [Poulida, 1996; Randel and Park, 2006; Li et al., 2005; Park et al., 2007; Ricaud et al., 90 2007; Ziemke et al., 2009] (ii) determining the height of convective outflow [Gettelman 91 and Forester, 2002; Froyd et al., 2009 (iii) identifying cases where outflow penetrates 92 the tropopause, into the LS [Hegglin, 2004; Ray et al., 2004; Liu and Zipser, 2005; 93 Chaboureau et al., 2007] (iv) comparing post-convective UT values measured by 94 satellite to model predictions [Liu et al., 2003; Halland et al., 2009; Mari et al., 2000] 95 and (v) the diurnal cycle of convection [Tian et al., 2004; Liu and Zipser, 2009]. 96

One result which is especially pertinent to this study is the trimodal distribution of tropical clouds, which can be traced to levels of convective detrainment. *Johnson et al.* [1999] reported clouds forming at three main levels, as detected by shipborne radar during the TOGA COARE experiment: low cumulus near 2 km and cumulonimbus at 15 to 16 km were most prominent, but cumulus congestus at approximately 5 km were frequent as well. These levels correspond well with stable layers associated with the trade wind inversion, the tropopause, and the 0°C melting level isotherm. Such stable levels are understood to be preferential regions of convective outflow (e.g. *Zuidema* [1998]). The mid-level cloud layer has also been documented by *Dessler et al.* [2006] using GLAS data, and by *Haynes and Stephens* [2007] and *Riley and Mapes* [2009] with CloudSat data. We will examine trace gas concentrations at the middle and upper levels of the trimodal distribution, to demonstrate the effects of convective transport.

The goal of this study is to use satellite measurements of CO combined with satellite-based locations of convective systems to quantify the impact of the convective systems on trace gas distributions. This study uses a 3 year archive of satellite data to search for evidence of convective transport of CO from the BL to mid and upper tropospheric levels. Observations are restricted to the tropics, where incoming solar radiation is strongest and convection is frequent. We use CO retrievals from the Tropospheric Emission Sounder (TES) on EOS Aura to detect mid-level concentrations, and CO retrievals from the Microwave Limb Sounder (MLS) aboard Aura for upper tropospheric values. To identify areas of convective transport we use estimates of rain rate derived

from microwave and IR satellite measurements, calibrated with Tropical Rainfall Measuring Mission (TRMM) observations, as a proxy for convective strength.

Carbon monoxide is a product of fossil fuel combustion and biomass burning, as well as

the oxidation of hydrocarbons, particularly methane [Jacob, 1999]. As such, its highest concentrations are near the Earth's surface. CO is lost in the troposphere primarily by reaction with the hydroxyl radical OH. The lifetime of CO generally ranges between one and two months, depending on season and location within the atmosphere. Carbon monoxide is not soluble and therefore not subject to rainout. These qualities make it an excellent tracer. CO is monitored for many reasons, among which: it is a significant anthropogenic pollutant, it is a precursor to O₃ in the troposphere, and it is a sink for OH (thus modulating the oxidation of many greenhouse gases). Global CO measurements are currently provided by a number of space-borne instruments, including Measurements Of Pollution In The Troposphere (MOPITT) [Drummond, 1992] aboard the EOS Terra satellite, Atmospheric Infrared Sounder (AIRS) [McMillan et al., 2005] aboard EOS Aqua, MLS [Waters et al., 2006] and TES [Beer et al., 2001], both on EOS Aura. We chose to use TES and MLS in this study: TES for its sensitivity in the mid-troposphere and its small footprint, allowing for unobscured profiles near or between clouds; MLS for its sensitivity near the tropopause, its capability to see through clouds, and its consistent proximity to TES profiles (as compared with MOPITT).

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To locate convection, one must detect an accompanying phenomenon, such as clouds resulting from condensing of lofted water vapor, or rainfall from those clouds. Numerous

approaches to cloud characterization exist, including diagnosing outgoing longwave radiation (OLR) [Liebmann and Smith, 1996; Lee et al., 2007], GOES Precipitation Index (GPI) [Arkin and Meisner, 1987] based on microwave brightness temperatures, and International Satellite Cloud Climatology Project (ISCCP) cloud classification [Rossow and Schiffer, 1991] based on cloud top pressure and optical depth. Techniques such as OLR and ISCCP yield high resolution products, but global coverage is only once or twice a day. GPI and its derivatives use data from geosynchronous satellites for good spatial and temporal resolution, but features such as high cirrus and non-raining portions of cloud anvils pose difficulties.

Microwave instruments that can detect precipitation beneath clouds are quite effective at locating convective rainfall. Successful estimation of rain rate from passive microwave satellite data was first demonstrated with ESMR on Nimbus satellites [Wilheit et al., 1973]. Current sensors include SSM/I on DMSP spacecraft, and TMI and PR aboard TRMM. Product resolution is quite good, but coverage is incomplete.

Algorithms such as the TRMM 3B-42 (see section 2.2) estimate rain rates with microwave data where available and calibrated IR data elsewhere, in addition to rain gauge values. This provides full coverage in the low and mid latitudes, every 3 hours, at quarter degree resolution. We presume that higher rain rates from the TRMM 3B-42 product imply greater likelihood of convection.

2. Approach

The coupling of the BL and FT by convection should have a definite signature in the trace gases. CO should appear aloft following a convective event that ventilates the BL, and outflow levels in the mid and upper troposphere should both show evidence of this lofting transport, since detrainment preferentially occurs near the stable layers of 0°C (occurring between 500 and 600 hPa) and the tropopause (~200 hPa). Thus there should be higher correlation between trace gas concentrations at mid and upper tropospheric levels following periods of local convective activity.

Our approach is to examine correlations between concentrations of CO in the middle and upper troposphere, and relate them to proximity to significant rainfall. First we examine this relationship within the context of a numerical model, the Global Modeling Initiative (GMI) chemistry and transport model (CTM). Using this model will allow us to examine what might be observed at the model scale of 2.5° longitude x 2° latitude. Then we search for a similar relationship using data from TES and MLS on Aura. We use TES pressure level 562 hPa and MLS level 215 hPa, because these correspond well with convective detrainment levels put forth in the trimodal cloud characterization [*Johnson et al.*, 1999]. We use the TRMM-calibrated 3B-42 rain rate product to identify likely instances of convection.

One challenge of using TES observations is that moist convection is accompanied by clouds, which prevent infrared-based remote sensing of the lower atmosphere. We anticipated however that cloud-free observations of the middle troposphere would be available immediately adjacent to convective areas, e.g. between clouds or along the

edges of convective storms. The small TES footprint makes this approach reasonable. We anticipate only a small fraction of the ~650 measurements made in the tropics (18°N-S) during each global survey will meet our criteria for cloud amount and proximity to convection. We compensate for infrequency by drawing upon a multi-year archive.

The remainder of the paper is structured as follows: The models (CTM and AGCM) are discussed in section 2.1; the data sources for CO and rain rate are presented in 2.2.

Results are given in section 3, and conclusions follow in section 4.

2.1 GMI and GEOS-4

NASA's Global Modeling Initiative has developed a 'Combo' CTM that combines photochemical schemes that are appropriate for the stratosphere and troposphere. The Combo CTM may be driven by meteorological fields from a general circulation model (GCM) or from a data assimilation system (DAS). GMI Combo includes gas phase reactions, photolysis, wet scavenging, dry deposition, heterogeneous chemistry (PSCs) and emissions [*Strahan et al.*, 2007; *Duncan et al.*, 2007]. For this study we have used the "Aura4" GMI experiment, run at 2° latitude by 2.5° longitude, with 55 vertical levels (hybrid sigma-pressure) up to 0.015 hPa. We use simulated CO that is recorded at times corresponding to the Aura afternoon overpass times, between 1 and 2 p.m. local time to best match the observation times of TES on Aura.

For the Aura4 experiment, GMI 'Combo' used meteorological fields taken from the Goddard Earth Observing System Data Assimilation System (GEOS DAS) version 4 [Bloom et al., 2005]. The GEOS DAS fields are produced at 1° latitude by 1.25° longitude on 55 levels; the fields are adapted for use at 2° by 2.5° in the CTM. The physics parameterization scheme consists of parameterizations for precipitation, clouds/radiation, turbulent mixing, and surface processes. The deep moist convection scheme is that of *Zhang and McFarlane* [1995], which uses a plume ensemble approach with convective updrafts and related downdrafts. Shallow convection is covered by the *Hack* [1994] scheme.

GEOS-4 DAS assimilates observations from many sources, including SSM/I (total precipitable water), MODIS (winds), geosynchronous satellites (winds), TOVS sounders (temperature and moisture), ERS-2 and QuikScat (sea-level winds), along with rawinsondes (temperature, winds, moisture), and surface stations (temperature, pressure, winds). The DAS outputs dozens of fields, including those that GMI requires, such as winds, temperature, precipitation and convective mass flux. Fields are saved at three hour intervals, in instantaneous and time-averaged forms; GMI 'Combo' uses the latter, horizontally averaged to 2° x 2.5°. In order to identify grid boxes with convective activity, we used the GEOS-4 DAS output field "PRECON", which stores the convective precipitation rate. Each GEOS-4 output file covers a single synoptic time period. In order to match the local ascending Aura overpass times implicit in each GMI output file, the convective precipitation field had to be extracted from multiple GEOS-4 files covering 24 hours of simulation.

2.2 TES, MLS and TRMM (3B-42)

The Tropospheric Emission Spectrometer (TES) is a Fourier transform spectrometer [Beer et al., 2001; Beer, 2006]. It is one of several instruments aboard NASA's Aura satellite, positioned in an ascending sun synchronous orbit with equator crossing at about 13:45 local solar time. TES measures in the infrared to detect trace gases including CO, water vapor, O₃ and methane. This study uses nadir profiles acquired during TES Global Surveys. Nadir view operates with a 5.3 x 8.3 km footprint, and each profile in a Level 2 file consists of data on 65 pressure levels ranging from nominal surface pressure to 0.1 hPa. Compared to the GMI model, TES has a much higher horizontal resolution, but of course the TES observations are much sparser than the GMI output (e.g. 9800 tropical profiles in July 2007, compared with 84,816 GMI gridbox values). And in terms of vertical resolution, the model levels within a grid column are independent compared to low vertical resolution retrieved TES profiles.

In the tropics, a TES tropospheric CO profile typically has between 1 and 2 degrees of freedom (DOF), a measure of the number of "independent" pieces of information that contributed to that profile. These pieces of information are used to modify the a priori (first guess) trace gas profile. (See *Beer* [2006] for an overview.) With a greater number of DOFs it is possible to achieve more variation from the a priori. In December 2005, TES implemented an Optical Bench warmup which resulted in better instrument

255 alignment, and average DOF values for TES CO measurements were significantly 256 improved. For this reason, we have only used data from January 2006 forward. 257 As a consequence of variations in TES DOF, some measurements reflect the actual state 258 259 of the atmosphere, while others rely heavily on the a priori. To distinguish between such 260 cases we consult the averaging kernel diagonal (AKD) values; where these are high 261 (close to 0.1), there is greater confidence that the true atmospheric values are retrieved 262 [Halland et al., 2009]. We tested several AKD thresholds and use an AKD threshold of 263 0.085 for most of this paper. In the final figure we show results using two other 264 thresholds for the AKD. 265 266 TES profiles used in this study were Level 2, Version 4. They were filtered using the 267 quality control fields, as in Table 5-3 of the EOS TES L2 Data User's Guide [Osterman et al., 2008]. In addition, since the TES IR-based retrievals are subject to interference by 268 269 clouds, profiles with nominal cloud top pressures less than 572 hPa were discarded. We 270 used profiles between 18°N and 18°S because DOFs changed abruptly beyond those 271 boundaries due to changes in the a priori. 272 273 The MLS instrument aboard Aura measures profiles of radiance from the Earth's limb 274 from upper troposphere to the mesosphere [Waters, 2006]. MLS scans columns ahead of 275 the satellite, every 165km along its orbit. MLS measures CO [Livesey et al., 2006; 276 Pumphrey et al., 2007], along with a dozen other atmospheric trace gases. As a

microwave instrument, it can detect constituents despite the presence of aerosols, cirrus or polar stratospheric clouds.

In this study we use CO as reported by MLS (version 2.2) at 215 hPa. At this level, the CO morphology and variability are adequately measured, but the values are biased by approximately a factor of 2 [*Livesey et al.*, 2008]. We compensate by dividing all values in half; we also processed the data according to guidelines in the MLS data quality document [*Livesey et al.*, 2007]. Each MLS CO profile was paired with the closest TES profile; the average separation was 195 km, and the average time difference was approximately 7 minutes.

For rain rate data, we turned to the 3B-42 product from the Tropical Rainfall Measuring Mission (TRMM) [Simpson et al., 1996]. This mission, a joint effort of the United States and Japan, was launched in 1997. It supports several instruments including the TRMM Microwave Imager (TMI), the Precipitation Radar (PR), and the Visible and Infrared Scanner (VIRS) [Kummerow et al., 1998]. The 3B-42 algorithm [Huffman et al., 2007] combines observations from several satellites, including TRMM, with rain gauge analyses. Computation of the multi-satellite portion begins with a merged microwave product including TMI and SSM/I rain rate estimates. Then VIRS data is used to produce an infrared/rain rate relationship, which is applied to geosynchronous IR data, such as that from the Geostationary Operational Environmental Satellite (GOES) series. The microwave products are high resolution (e.g. TMI is 5.1 km), but measurements accumulated over timescales of up to three hours do not completely cover the tropics.

The IR measurements are similar resolution (e.g. GOES is 4 km), and a merged product from several geo-IR satellites provides excellent coverage (60° N-S) at half hour intervals [*Janowiak et al.*, 2001]. Incorporating both microwave and IR data, the 3B-42 rain rate product is generated for latitudes covered by TRMM (50° N-S) at quarter degree resolution and 3 hour intervals.

3B-42 does not distinguish between convective rain, which is linked to strong vertical transport, and stratiform rain, which is not. However, it is generally true that greater rain rates are more likely associated with convective systems. Indeed, the ratio of convective to stratiform rain rates in the tropics is 4.1, when measured at PR resolution [Schumacher and Houze, 2003]. So the error introduced by the inclusion of stratiform rain rates in this study should be relatively small.

Rain rates were used as follows: For a given pair of TES and MLS CO profiles we searched a neighborhood of 3B-42 grid boxes centered on the spot midway between the profiles. Since 3B-42 rain rates are available every three hours, we used the values from the 3-hour window prior to the TES observation, to ensure that we would see the effects of vertical mixing.

319 3. Results

321 3.1 GMI and GEOS-4

We first analyze simulated upper and mid-tropospheric CO to demonstrate the expected relationship between composition and convective transport as implemented in the GMI CTM.

3.1.1 Scatter plots varying rain rate threshold

Daily GMI CO concentrations from July 2007 were compared at two different pressure levels, 562 hPa and 215 hPa, selected to match the TES and MLS pressure levels used in the next section. The 562 hPa level represents the level of expected convective detrainment around the 0°C isotherm, while 215 hPa corresponds with outflow near the tropopause. As shown in the first scatter plot of Figure 1, the majority of CO values on both pressure levels range between 50 and 150 ppbv. Panel (a) includes values from all GMI grid boxes in the tropics, 19°S - 19°N, regardless of rain rate. The broad pattern matches our expectations: Points along the diagonal are consistent with vertical convective transport to both levels of the troposphere; areas with substantial CO at lower altitudes but minimal amounts above may represent cases where shallow convection has deposited CO emissions from the boundary layer up into the mid-troposphere. Areas with greater CO concentrations at 215 hPa may represent cases where deep convection has been dominant. The off-diagonal cases could also be due to the effects of horizontal advection transporting CO from other convective areas.

Figure 1, panels (b)-(d) show subsets of the original data, where profiles were discarded if the convective rain rate was less than 0.5, 1.0 and 2.0 mm/hr respectively. As

convective activity increases from one panel to the next, profiles with markedly different CO concentrations at the two levels are eliminated. This fits with our hypothesis, in that convective transport is coupling the middle and upper levels of the troposphere. In the last panel, where convective rain rates are highest, the remaining points are clustered along the diagonal, and the correlation coefficient is quite high, at 0.967. This simple test demonstrates that the GMI implementation of trace gas transport, as driven by GEOS-4 winds and convective mass flux, is consistent with our conceptual model of the effect vertical convective mixing will have on the trace gas distributions. This is an idealized case with which to compare the TES/MLS observations.

3.1.2 Geographic Location of GEOS-4 Rain

Another useful plot for comparing the model and satellite results is the geographic location of dominant rainfall. The areas of convective rain in the GEOS-4 DAS are shown in Figure 2. The grid boxes are shaded based on the number of days that met or exceeded the rain rate threshold. Note the prominence of the ITCZ in panels (a) and (b), and the areas of most frequent convective rain such as Costa Rica and north of Papua New Guinea, in panel (c). These plots confirm the success of GEOS-4 DAS in reproducing realistic rain distribution patterns.

3.2 TES, MLS and TRMM (3B-42)

The TES and MLS observations are quite sparse in comparison to model output. Within our constraints of tropical latitudes and sufficiently cloud-free observations, a month of TES Global Survey data yields roughly 30 to 60 profiles. Filtering these using rain rate thresholds quickly reduces the sample size to numbers too small for a meaningful analysis. So we chose to include three full years of TES and MLS data in this study, from 2006 through 2008.

In order to estimate the strength of convection in the neighborhood of TES and MLS measurements, we consulted the TRMM-calibrated rain rate data for the 3-hour interval prior to the TES overpass time. Our initial tests included an area similar in size to a GMI gridbox: we searched an 11x7 patch of TRMM values centered between the TES and MLS profiles, and took the maximum rain rate. (Since TRMM 3B-42 is 0.25° resolution, the area was 2.75° in longitude by 1.75° in latitude.) In addition we tested our approach with smaller patches of TRMM grid boxes, including 7x5, 5x3 and 3x3. As the area in which we searched for rain grew smaller, the effects of the vertical mixing became more apparent. So we begin by examining the 3x3 case, wherein each patch is a square of 9 grid boxes; the maximum rain rate is sought within a 0.75° x 0.75° region.

3.2.1 Scatter plots varying rain rate thresholds

Plots of CO correlation between pressure levels 562 and 215 hPa are shown in Figure 3.

Panel (a) includes all qualifying profiles, regardless of local rain rate. It shows that the majority of values at midlevel are between 50 and 150 ppbv, while the range on the upper

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level is between 50 and 125 ppbv. The (solid) line of linear fit shows that greater CO concentrations exist at lower altitudes. This can be attributed to diffusive transport from the boundary layer, particularly during times of weak trade inversion. It may also be due in part to the approximate correction we have applied to compensate for the MLS high bias. Greater CO concentrations at mid-level were also present in GMI (see the line of linear fit in Figure 1), but in the TES/MLS case this is much more pronounced. Panels (b)-(d) in Figure 3 show the relationship between middle and upper tropospheric CO as the rain rate requirement for the TES/MLS observation neighborhood is increased. As the threshold is increased from 1 up to 3 mm/hr, the correlation steadily increases as well. TES/MLS values associated with higher rain rates are clustered around the diagonal, as was the case in the GMI plots (Figure 1). Clustering is tighter for GMI, because values are reported immediately after the convective mass flux is applied to the trace gas field, as opposed to the TES/MLS measurements which were taken up to 6 hours after rainfall. The interval between convective rain and CO observations gives advection time to alter distributions. Part of the difficulty in characterizing patterns in this data is the remarkably low number of profiles satisfying our criteria, as shown in Table 1. This is due in part to the rejection

of profiles satisfying our criteria, as shown in Table 1. This is due in part to the rejection of cloudy TES profiles. For the 0 mm/hr case, 84% of the profiles were filtered out based on their cloud top pressure. In the case of 3 mm/hr, 90% were removed. The linkage of additional clouds with higher rain rate was expected, since convective rain is

accompanied by updrafts, leading to detrainment, condensation, and therefore clouds.

To establish a level of confidence in the correlation values computed for small sample sizes, we employed a bootstrap approach: we computed the correlation for a randomly chosen subset of half the points, repeated this process 1000 times, and made histograms of the results. We accepted the correlation over the full sample only when it matched the average of the bootstrap correlations to within 0.05, and when the central 90% of the bootstrap values did not exceed a range of 0.5. Each of the correlations listed in Table 1 satisfied these criteria. The bootstrap histograms are shown as insets in Figure 3, with markers denoting the 5%, 95% and average values.

3.2.2 Geographic Location of TES/MLS Observations

Because strong convection is usually accompanied by substantial clouds, we anticipated that non-cloudy TES profiles linked with higher rain rates would be adjacent to regions of heavy rain, rather than within such regions. The locations of accepted TES/MLS profile pairs are shown in Figure 4. In panel (a), rain rates are not considered, and the pattern arises solely from the elimination of cloudy profiles. The areas of highest number of observations are those that tend to have low levels of rain, such as Saudi Arabia and the horn of Africa. The fewest profiles are found in areas of high rainfall, such as the ITCZ, Maritime Continent, Indian Ocean and Brazil. So, as expected, most non-cloudy TES observations are in the areas of least convective rain. Once a rain rate threshold is applied a different pattern emerges. (Note that the color scale for panels (b) and (c) is different from that in the first panel.) As anticipated, the profiles in rainy neighborhoods

tend to be located adjacent to areas of frequent rain, such as the ITCZ, Brazil, and the Indian Ocean. We interpret these to be instances where TES measures CO in sufficiently cloud-free columns along the edges of convective storm systems. The convective lofting of CO by such systems is evident in the TES and MLS data. With higher rain rates there is greater likelihood of convective activity in the area, and therefore greater vertical mixing of CO and higher correlation between levels.

3.2.3 Correlation affected by size of TRMM neighborhood, and AK Diagonal

How does the correlation between mid and upper levels of CO change as the distance from convection increases? And how is the correlation affected by our choice of AKD threshold? To answer these questions, we performed two sensitivity experiments. First we varied the size of the search neighborhood in the TRMM-based rain rate data, testing four different patch sizes: 11x7, 7x5, 5x3 and 3x3 grid boxes, in the 0.25° x 0.25° grid. Larger search areas allow for rain rate maxima to be identified at greater distances from the CO profiles. In the second sensitivity test we varied the minimum TES AKD accepted: 0.08, 0.085 and 0.09. This was done to determine the trend in correlations as TES values gained greater independence from the a priori and more accurately reflected the state of the atmosphere. The results from both of these experiments are shown in Figure 5. Multiple patch sizes are plotted in each panel, and a different minimum AKD applies to each panel. (Note that the original series of correlations, recorded in Table 1, appears as the sequence of red asterisks in the middle panel.)

460 The influence of TRMM search area size is evident in panel (a). Each gray square 461 designates the CO correlation derived from TES/MLS profile pairs close to a sufficiently 462 high TRMM rain rate, determined by searching an 11x7 neighborhood of TRMM values 463 (2.75° x 1.25°). The blue triangles show the results from searching smaller 464 neighborhoods of 7x5 grid boxes (1.75° x 1.25°). The green diamonds reflect 5x3 patches 465 $(1.25^{\circ} \times 0.75^{\circ})$, and the red asterisks use 3x3 patches $(0.75^{\circ} \times 0.75^{\circ})$. As the search area is constricted, the average distance between the TES/MLS profiles and 466 467 maximum rain rate is reduced, so the lofting effects of convection should be more evident 468 in the CO measurements, and we anticipated higher correlation as a result. This is clearly 469 exhibited in the plots; while the 11x7 sequence does not show an increase in correlation 470 as local maximum rain rate increases, the 7x5 sequence does show this relationship, and 471 the 5x3 and 3x3 cases even more so. This demonstrates that trace gas correlation between 472 the middle and upper troposphere depends on proximity to convection as well as 473 convective intensity. 474 475 The series of three panels shows the impact of more stringent TES AKD requirements. 476 Two effects are apparent. First, as average AKD increases, the CO correlation for a given 477 rain rate tends to increase as well. This means that the trends we are seeing derive 478 primarily from the atmospheric state, and should not be attributed to the TES a priori. 479 The second effect is the reduction in qualified TES profiles, resulting in smaller sample 480 sizes over which to evaluate correlations. This paucity of data is further exacerbated 481 when profiles are filtered based on rain rate. With TES AKD \geq 0.08, before rain rate is 482 considered, there are 17,358 profile pairs from the 2006-2008 period; with AKD \geq 0.085

there are 9,945 pairs, and with AKD \geq 0.09 only 5,091 pairs qualify. Filtering these sets with increasing rain rate thresholds quickly reduces the sample sizes to fewer than 50, and soon thereafter the bootstrap confidence test (see section 3.2.1) begins to fail, discontinuing each sequence. For example, the 3x3 sequence in panel (c) is computed for rain rates at or above 0, 0.5 and 1.0 mm/hr, using 5,091, 83 and 41 profile pairs respectively. At 1.5 mm/hr, too few profiles remain to calculate a statistically meaningful correlation. We assert that, despite the short series of correlations plotted in the last panel, they confirm the trends seen in the prior two panels, namely that the proximity and intensity of convection influences the degree of correlation in middle and upper tropospheric concentrations of CO.

4. Conclusion

Using rain activity as a proxy for convection we have demonstrated a link between tropical convection and CO concentrations in the middle and upper troposphere, at levels corresponding to the stable layers at 0°C and the tropopause. In other words, convective activity is pumping CO from the surface into the mid and upper troposphere – chemically coupling the tropospheric layers. This association was first demonstrated in the GMI CTM. Correlations between CO concentrations at 562 and 215 hPa increased as the convective precipitation rate increased. This conformed to our expectations. In addition, the locations of high convective rain rates for the month we studied aligned well with standard patterns of tropical rain, e.g. the ITCZ.

For the observational study, CO measurements from TES at 562 hPa (unobscured by clouds) and from MLS at 215 hPa were examined, together with rain rates from the TRMM 3B-42 product. First we examined the influence of rain rate on the correlation, using a 0.75 x 0.75 search area within TRMM data for each pair of TES/MLS profiles, and using TES profiles with at least an AKD value of 0.085. As the rain rate threshold was increased, the correlation between CO burdens at the two levels grew stronger, as we anticipated; greater convective vertical mixing is concomitant with heavier rain. However, the sample size (profile count) fell dramatically as we restricted ourselves to higher local rain rates. This was because rain is accompanied by clouds (convective rain leads to additional cloud formation), and we discarded TES profiles that were not sufficiently cloud-free.

The influence of clouds was also evident in the location of qualifying profiles. Most profiles clustered in areas of minimal rain, such as the horn of Africa. When rain rate thresholds were enforced, the remaining profiles appeared at the periphery of major cloudy areas such as the ITCZ and Brazil. These plots differed greatly from the GMI location plots, principally due to the cloud-free condition we imposed on TES profiles. Fortunately, the small footprint of TES permitted some measurements not obscured by clouds immediately adjacent to regions with rain. We extended the study to include 36 months of data to provide a sufficient number of non-cloudy profiles. Also we used a bootstrap statistical method to gauge confidence in cases with only a small number of CO measurements. By these means we were able to show linkage between rain rate and CO burden correlation similar to that shown with the model.

We explored the sensitivity of this result to two parameters: the rain search area, and the TES averaging kernel diagonal values. For the former, we varied the extent of search within TRMM 3B-42 for a maximum rain rate to associate with each TES/MLS profile pair. TRMM data neighborhoods of four sizes, ranging from 2.75°x1.75° (approximately the size of one GMI gridbox), to 0.75°x0.75° were tested. For each areal size, our first result was reproduced: increased rain rates led to higher correlations. Smaller search domains also tended to yield better correlations, particularly for greater rain rates. This result is consistent with the idea that closer proximity to convective mixing translates into a stronger observable signal. When we required higher AKD values for TES, the correlations again tended to rise, giving us confidence that the signal in CO reflected actual atmospheric concentrations, rather than strong influence from the a priori. We showed significant correlations up to AKD threshold 0.09, beyond which the profile count was too small for meaningful results.

We have demonstrated that tropical convection plays an important role in transporting trace gases vertically within the troposphere, with effects in keeping with the trimodal characterization of tropical clouds, in which boundary layer parcels are convectively lofted to stable levels in the troposphere. Satellite observations show that this process operates on a wide scale throughout the tropical regions where strong convection is prevalent.

551	In this paper we examined how correlation was affected by proximity to rainfall but
552	subsequent advection of the CO plume was not considered. A follow-up study could
553	focus on cases where convection was located upwind of the CO measurement.
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776 Figure Captions

Figure 1. GMI CO concentrations at middle and upper tropospheric levels, using one month of daily output between 19°S-19°N. Panel (a) shows values from all grid boxes. In subsequent panels, values from a grid box are only included if the convective precipitation rate within that grid box meets the given threshold. The correlation of CO between the two levels increases as the rain rate increases.

Figure 2. Each grid box shows the number of days (day count) in July 2007 for which GEOS-4 computed a tropical convective rain rate satisfying the given threshold, at Aura afternoon overpass times. Areas of most frequent convective rain correspond with well-known areas of convection e.g. the ITCZ and the Maritime Continent. In panel (a), rain rate ≥ 0.05 mm/hr, panel (b) ≥ 0.5 mm/hr, and panel (c) ≥ 1 mm/hr.

Figure 3. CO concentrations at middle vs. upper tropospheric levels, as measured in the tropics (18°S-18°N) from Jan 2006 through Dec 2008. TES Global Survey data at 562 hPa were used, only where the Averaging Kernel diagonal was at least 0.085. MLS data were used at 215 hPa, with values halved to compensate for a factor of 2 bias at this level. The profiles were screened for quality and cloud presence as described in the text. Panel (a) shows values from all qualifying profiles. In subsequent panels, a profile is only included if the maximum local TRMM-based rain rate (within a 3x3 grid patch) meets the given threshold. As the 3B-42 rain rate threshold increases, so too does the correlation of CO between the two levels; at the same time, the number of points decreases pretty dramatically. Histogram insets show the distribution of 1000 bootstrap test correlations,

ranging from -1 to 1. Markers below the histograms show the 5%, average, and 95% locations; markers above show the correlation coefficients using all data.

Figure 4. Similar to Figure 2; however, this figure shows the geographic distribution of TES/MLS profile pairs with TRMM rain rates satisfying thresholds 0.0, 0.05 and 1.0 mm/hr (panels (a) and (c) correspond to panels (a) and (b) in Figure 3). The profiles here are binned at 2° x 2.5° to match Figure 2. In panel (a), no rain rate threshold is applied, and the cloud-limiting filter for TES leaves clusters of profiles in areas of very little rain (e.g. parts of Africa, Saudi Arabia and the South Atlantic). Panels (b) and (c) show the effects of imposing rain rate thresholds of 0.05 and 1.0 mm/hr; this greatly reduces the number of profiles, so a different color scale is used. As the required rain rate increases, the remaining profiles are frequently located near areas of prominent rain (e.g. the ITCZ, Maritime Continent, and Brazil).

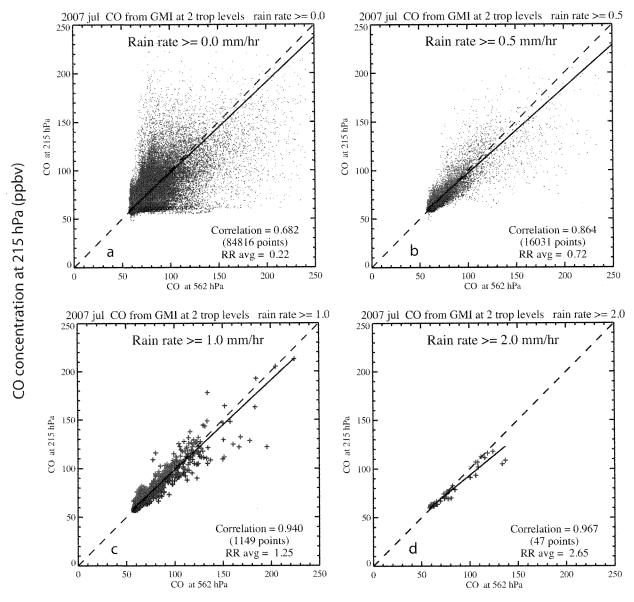
Figure 5. Correlation of CO measurements from TES (562 hPa) and MLS (215 hPa), plotted against rain rate thresholds satisfied by TRMM 3B-42. Four methods of rain rate determination are shown in each panel; they differ only in the number of TRMM values consulted in the vicinity of the TES/MLS observations, ranging from 77 (11x7) down to 9 (3x3). Each panel imposes a different AKD minimum for the TES data: 0.08, 0.085 and 0.09. Data spans three full years, 2006-2008. The points on the red/asterisk line (3x3 patches) in panel (b) correspond to the correlation coefficients shown in Table 1. Correlation is not recorded if it does not satisfy the bootstrap criteria described in section 3.2.1.

824 Tables

Rain Rate Threshold (mm/hr)	Θ	0.5	1.0	1.5	2.0	2.5	3.0
Correlation Coefficient	0.454	0.538	0.604	0.618	0.719	0.780	0.804
Profile count	9940	213	108	64	45	33	22

Table 1. Interlevel CO Correlation Coefficients

CO correlations between middle and upper tropospheric levels associated with various rain rate thresholds. Cases shown include those from Figure 3 along with intermediate rain thresholds.



CO concentration at 562 hPa (ppbv)

